
Algorithm 2: Masking routine for MLM

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function tokenize_for_mlm(batch)
  // 1) Build paired texts: word - desc [SEP] greek_word - greek_desc
  tokenized_texts ← [];
  foreach (w, d, gw, gd) in zip(batch[word], batch[description], batch[greek_word],
    batch[greek_description]) do
    full_text ← f"{w} - {d} [SEP] {gw} - {gd}";
    ids ← tokenizer.encode(full_text, add_special_tokens=True, max_length=512, truncation=True);
    append(tokenized_texts, ids);

  // 2) Pad to max length in batch
  input_ids_tensors ← [tensor(ids) for ids in tokenized_texts];
  input_ids_padded ← pad_sequence(input_ids_tensors, batch_first=True,
    padding_value=tokenizer.pad_token_id);
  // 3) Apply MLM masking (15%)
  (input_ids_masked, labels) ← mask_tokens(input_ids_padded, tokenizer, mlm_probability=0.15);
  // 4) Attention mask: 1 for non-pad
  attention_mask ← (input_ids_masked ≠ tokenizer.pad_token_id).long();
  return { "input_ids": input_ids_masked,;
    "attention_mask": attention_mask,;
    "labels": labels };

function mask_tokens(inputs, tokenizer, mlm_probability=0.15)
  // 1) Labels start as a copy of inputs
  labels ← clone(inputs);
  // 2) Sample candidate positions with Bernoulli(mlm_probability)
  probability_matrix ← full(shape(labels), mlm_probability);
  masked_indices ← bernoulli(probability_matrix).bool();
  // 3) Exclude special tokens: PAD, CLS, SEP
  special_mask ← zeros_like(labels, dtype=bool);
  foreach tok ∈ {pad_id, cls_id, sep_id} do
    special_mask ← special_mask ∨ (labels == tok);
  masked_indices ← masked_indices & not(special_mask);
  // 4) Non-masked positions set to -100 in labels (ignored by loss)
  labels[not(masked_indices)] ← -100;
  // 5) Replace 80% of masked positions with [MASK]
  replace_prob ← full(shape(labels), 0.8);
  indices_replaced ← bernoulli(replace_prob).bool() & masked_indices;
  inputs[indices_replaced] ← tokenizer.mask_token_id;
  return (inputs, labels);
```
